



# Modelling the impact of new parkrun events: efficiency vs equity

R for the Rest of Us Podcast

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Dr Robert Smith | January 2025



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<https://github.com/dark-peak-analytics>



<https://www.linkedin.com/company/dark-peak-analytics>



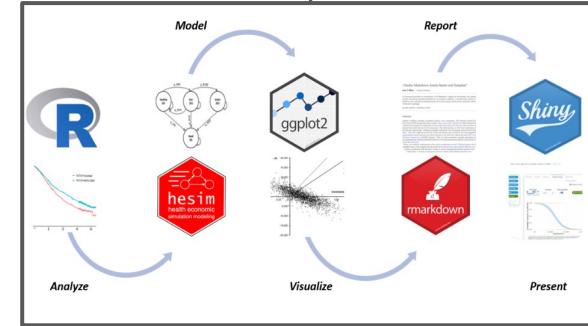
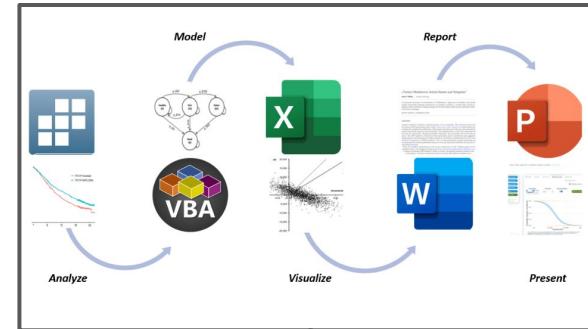
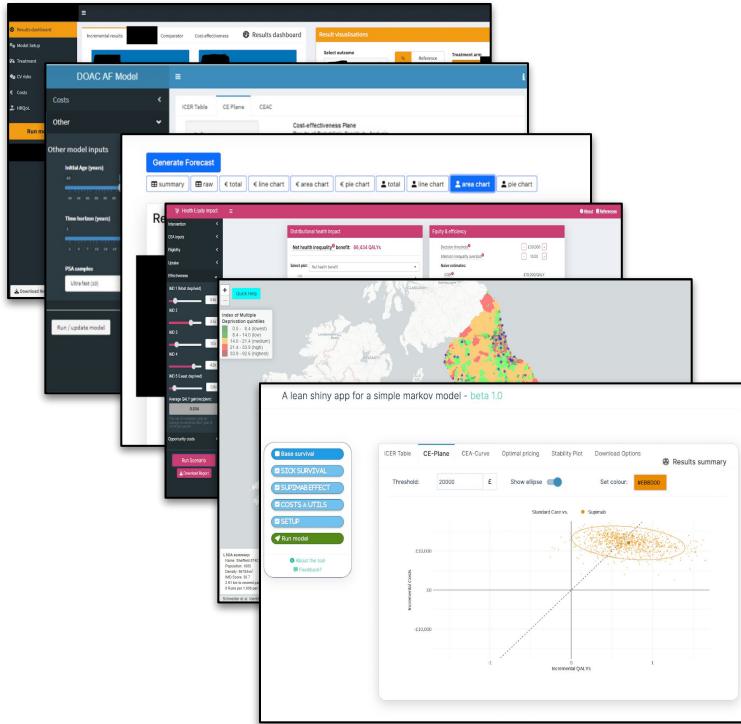
# Who am I?



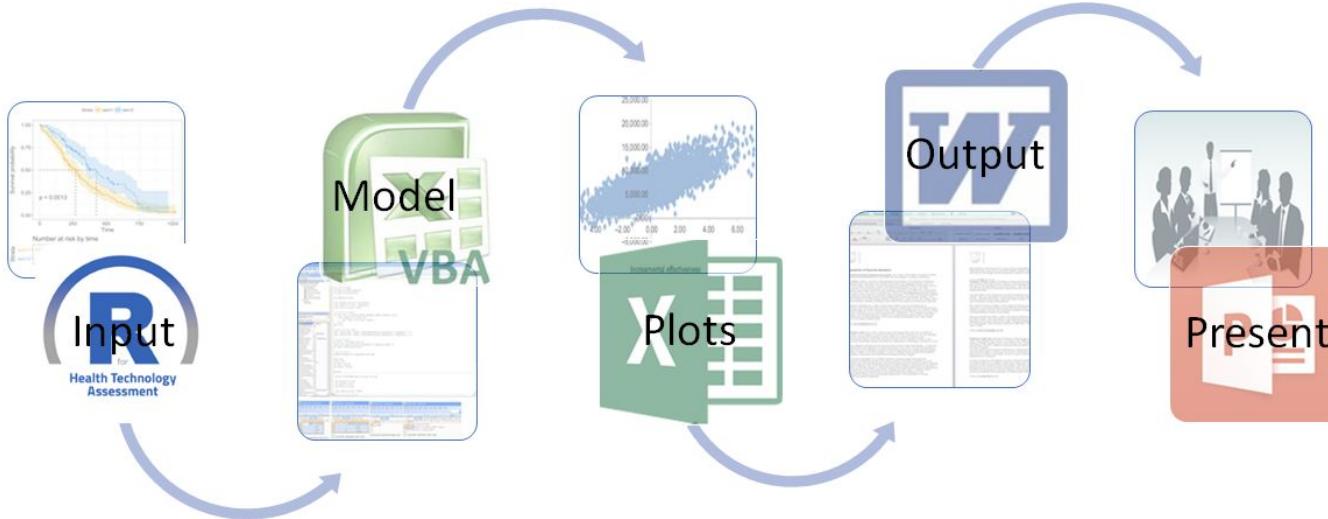
Robert Smith, PhD

Affiliations: Dark Peak Analytics, University of Sheffield

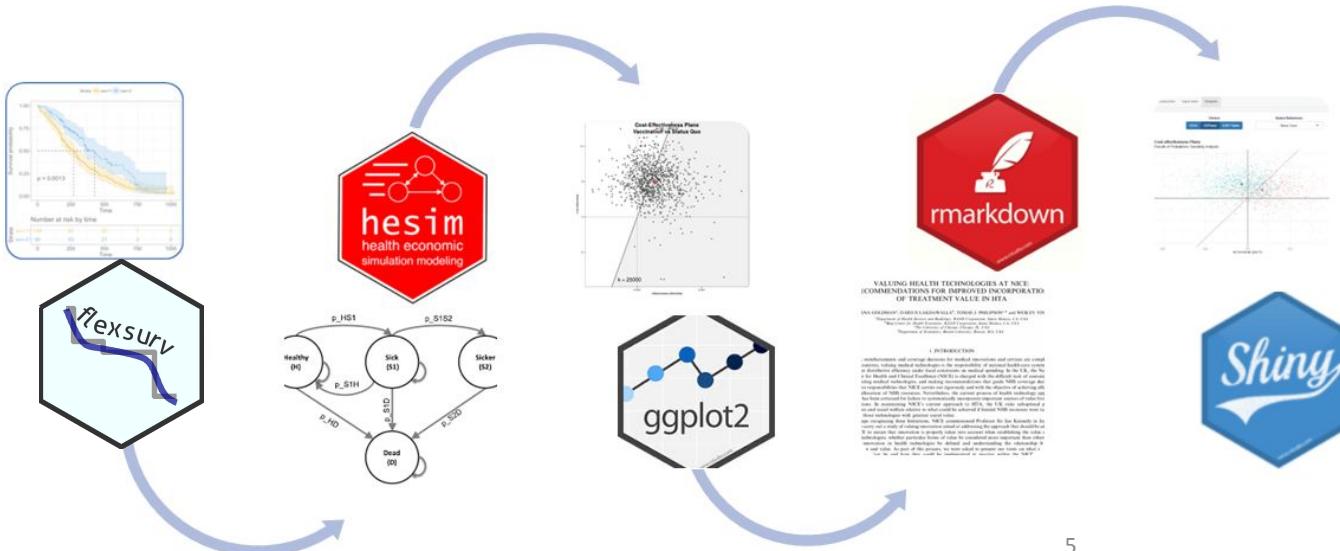
# Shifting Health Economics to Script Based Models



# Current process



# Future process





# User-interfaces

Wellcome Open Research

Wellcome Open Research 2020, 5:69 Last updated 05 July 2022

METHOD ARTICLE

**Making health economic models Shiny: A tutorial**

[version 2; peer review: 2 approved]

Robert A. Smith, Paul Schneider

School of Health and Related Research, University of Sheffield, Regents Court, Sheffield, S1 4DA, UK

Equal contributors

First published: 14 April 2020, 5:69  
Latest published: 31 Jul 2020, 5:69  
<https://doi.org/10.12688/wellcomeopenres.15807.2>

**Abstract**

Health economic evaluation models have traditionally been built in Microsoft Excel, but more sophisticated tools are increasingly being used as model complexity and computational requirements increase. Of all the programming languages, R is most popular amongst health economists. It is a free, open-source language that is easy to learn and is highly flexible. However, even with an integrated development environment such as R Studio, R lacks a simple point and click user interface and lacks the ability to programme graphical ability. This might mean the switch from Microsoft Excel to R seems daunting, and it might make it difficult to directly communicate results with decision makers and other stakeholders.

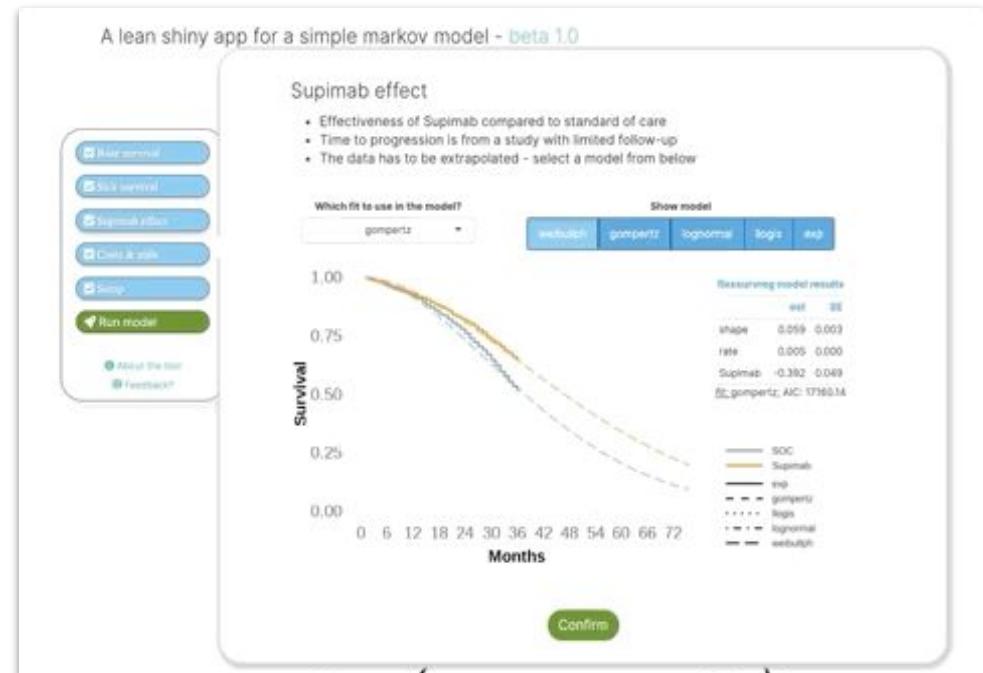
The R package Shiny has the potential to resolve this limitation. It allows programmes to be run in a web browser and can be developed in R and programmed with a lower level user interface. Users can specify their own assumptions about model parameters and run different scenario analyses, which, in the case of regular a Markov model, can be computed within seconds. This paper provides a brief introduction to Shiny and demonstrates how it can be used to build a shiny application. We use a four-state Markov model developed by the Decision Analysis in Health (DARTH) group as a case study to demonstrate main principles and basic functionality.

A more extensive tutorial, all code, and data are provided in a GitHub repository.

**Keywords**

Health Economics, R, RShiny, Decision Science

Page 1 of 26



Smith RA and Schneider PP. Making health economic models Shiny: A tutorial. Wellcome Open Res 2020, 5:69.  
<https://doi.org/10.12688/wellcomeopenres.15807.2>

<https://darkpeakanalytics.shinyapps.io/sadm-mk2/>



OK - stop talking about yourself!



# Research collaboration



The  
University  
Of  
Sheffield.



Dr. Paul Schneider



Dr. Robert Smith

This work was supported by funding from the Wellcome Trust [108903/B/15/Z and 108903] and the University of Sheffield. The funders had no role in study design; in the collection, analysis and interpretation of data; in the writing of the report; or in the decision to submit the article for publication.

# Background



<https://www.youtube.com/watch?v=ktx4lujARdc&feature=share>

# Background



# Background

## PARTNERSHIP WITH PARKRUN WORTH £3M

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**Collaboration aims to create 200 new events and boost participants from under-represented groups**



12 December 2018



News



Funding





# Ethnicity, Deprivation & Access

Wellcome Open Research Wellcome Open Research 2020, 5(9) Last updated: 05 July 2022



RESEARCH ARTICLE

 Does ethnic density influence community participation in mass participation physical activity events? The case of parkrun in England [version 2; peer review: 3 approved]

Robert A. Smith<sup>1</sup>, Paul P. Schneider<sup>2</sup>, Alice Bullas<sup>2</sup>, Steve Haake<sup>2</sup>, Helen Quirk<sup>2</sup>, Rami Cosulich<sup>1</sup>, Elizabeth Goyder<sup>1</sup>

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v2 First published: 16 Jan 2020, 5(9) <https://doi.org/10.31238/wellcomeresources.15657.1>

Last published: 18 Jun 2020, 5(9) <https://doi.org/10.31238/wellcomeresources.15657.2>

**Abstract**  
**Background:** parkrun has been successful in encouraging people in England to participate in their weekly 5km running and walking events. However, there is evidence of heterogeneity in parkrun participation across different communities in England. After controlling for travel distances, deprived communities have significantly lower participation rates.

**Methods:** This paper expands on previous findings by investigating deprivation and participation by ethnic density. We combined geo-spatial data available through the Office for National Statistics with participation data provided by parkrun, and fitted multivariable Poisson regression models to study the effect of ethnic density on participation.

**Results:** We find that areas with higher ethnic density have lower participation rates. This effect is independent of deprivation.

**Conclusions:** An opportunity exists for parkrun to engage with these communities and reduce potential barriers to participation.

**Keywords:** parkrun, Physical Activity, Ethnic Density, Deprivation

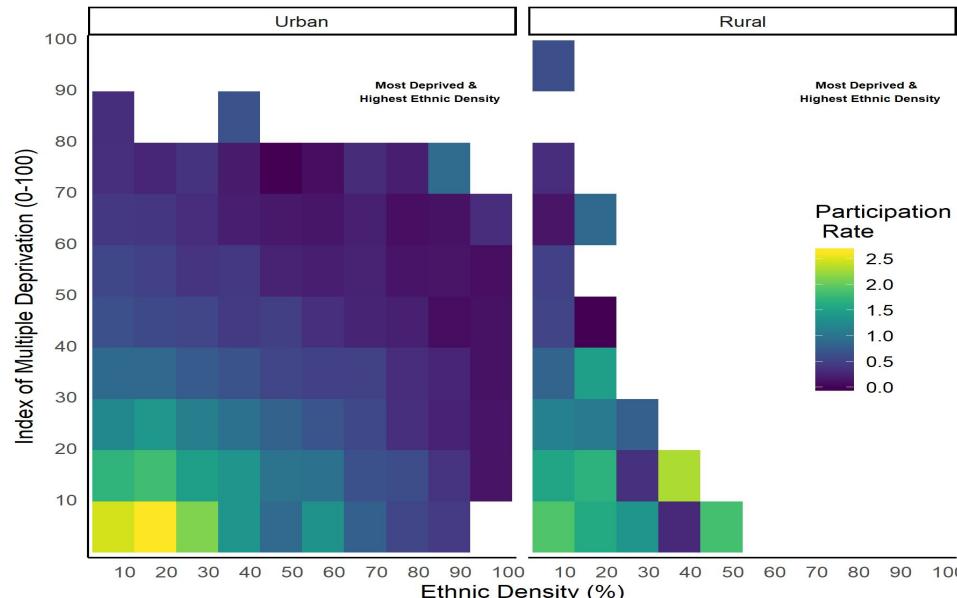
Open Peer Review Approval Status: ✓✓✓

	1	2	3
version 2	✓	✓	✓
(revision)	---	---	---
18 Jun 2020	---	---	---
version 1	✓	✓	?
16 Jan 2020	---	---	---

1. Anne Grunstein<sup>1</sup> The Australian Prevention Partnership Centre, NSW, Australia  
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 2. Goode Oakink<sup>1</sup> University of St Andrews, St Andrews, UK  
 3. Stephen Senn<sup>2</sup> Consultant Statistician, Edinburgh, UK

Any reports and responses or comments on the article can be found at the end of the article.

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Sources: Office for National Statistics and parkrunUK

# Longitudinal study - participation & access



Socioeconomic inequalities in distance to and participation in a community-based running and walking activity: A longitudinal ecological study of parkrun 2010 to 2019<sup>a,b,c,d,e</sup>

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## ARTICLE INFO

**Keywords:** Parkrun  
Physical activity  
Socioeconomic deprivation  
Ecological study  
Relative index of Deprivation

**Objectives:** To conduct a longitudinal ecological analysis of the distance to and participation in five weekly community physical activity events (parkrun) in England from 2010 to 2019, and related socioeconomic and ethnic inequalities, to inform policies to support participation in physically active community events.

**Methods:** Monthly data were collected from parkrun, the largest free-to-enter community running event in the United Kingdom (UK), each month from January 2010 to December 2019. We then report the trends in distance to and participation in parkrun by Index of Multiple Deprivation quintile. We also report trends in the Relative Index of Deprivation (RID) and the Index of Multiple Deprivation (IMD) quintiles. We used Poisson regression models to predict trends in IMD level determinants (e.g. deprivation and ethnicity) of parkrun participation between 2010 and 2019, using time series analysis.

**Results:** Mean distance to the nearest parkrun decreased from 34.3 km in 2010, to 4.6 km in 2019. Throughout the period, parkrun events tended to be situated closer to deprived areas compared to less deprived areas. The mean number of monthly parkrun participants increased from 1.1 million in 2010 to 1.8 million in 2019, showing a linear growth. Participation over the period exhibited a clear socioeconomic gradient, with people from deprived communities having consistently lower participation rates than the period, particularly those from more deprived areas between 2010 and 2013 (H11 Poisson regression model:  $\beta = 0.39$ ,  $P < 0.001$ ) and between 2012 and 2019 ( $\beta = 0.30$ ,  $P < 0.001$ ). The results of the Poisson regression model validate this finding, the coefficients on IMD quintiles were significant ( $\beta = 0.29$ ,  $P < 0.001$  and  $\beta = 0.20$ ,  $P < 0.001$  respectively) and the coefficient on ethnicity was non-significant ( $\beta = 0.025$ ,  $P = 0.42$ ).

**Conclusion:** Over the past 10 years, parkrun distance to the nearest parkrun decreased from a mean of 34 km to 5 km. In 2010, there was equality between the least and most deprived areas but by 2017 the distance of the most deprived areas was 29% that of the least deprived. Participation was shown to have increased over the past 10 years, with the rate of increase being higher in the most deprived areas. There was a significant socioeconomic inequality in participation (d) dramatically; from 2010 to 2019 participation increased linearly, and inequality in participation increased (d) exponentially. The results suggest that parkrun has been successful in reaching more deprived communities, the socioeconomic gradient in participation rates remained high and stable since 2013. Gaining a better understanding of the reasons why parkrun grew so quickly may be useful for other physical activity movements, while further analysis of the relatively lower participation rates in areas with higher

\* R.S., P.S. & R.C. are joint funded by the Wellcome Trust Doctoral Training Centre in Public Health, Economics and Decision Science (130902) and the University of Sheffield. H.Q. is funded by an NHR School of Public Health Research (SPHR) post-doctoral funding fellowship. The funders had no role in study design, data collection and interpretation, or writing of this manuscript. We would like to thank Steve Haake (Head of Analysis at parkrun) and Christopher Wellington (Global Head of Health and Wellbeing at parkrun) for providing area-level parkrun participation data.

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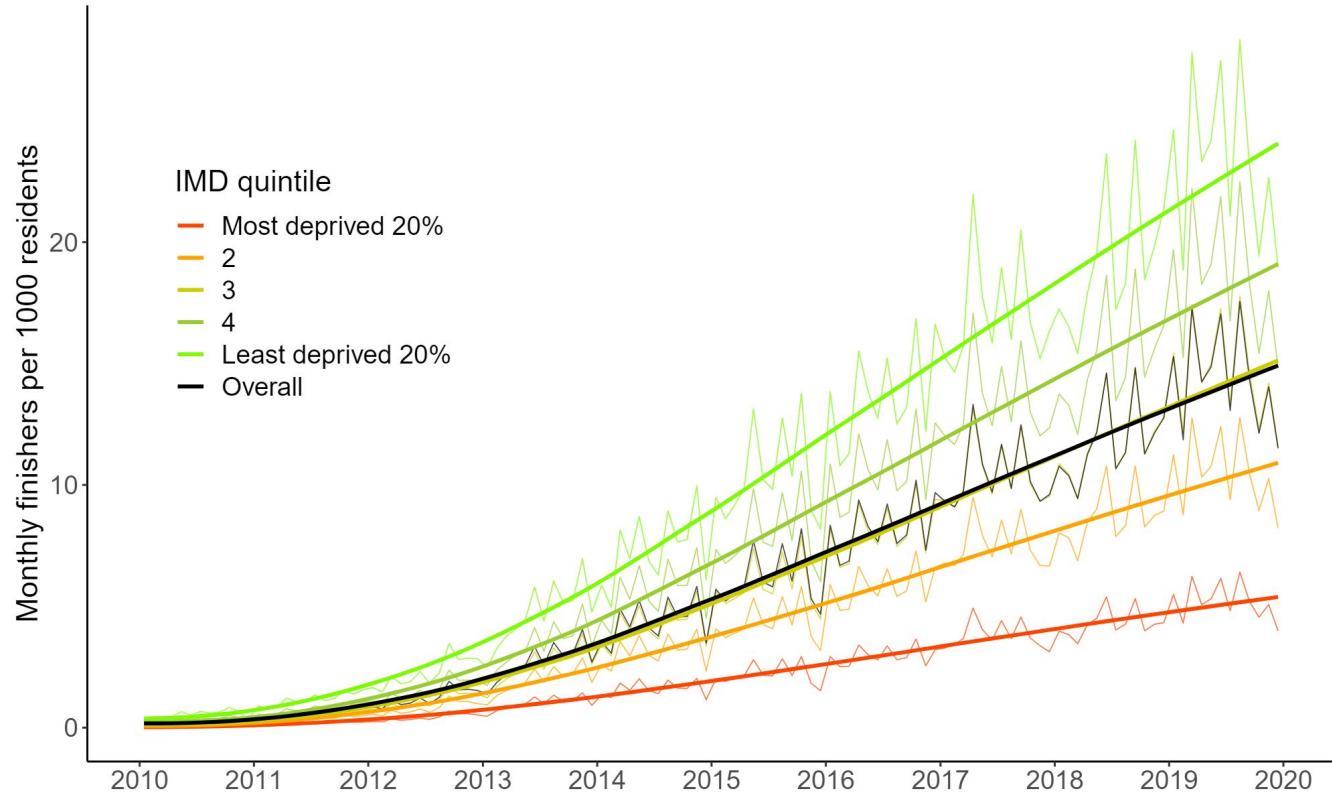
<sup>d</sup> Joint first author.

<sup>e</sup> Data available at <https://doi.org/10.1101/161026>.

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1253-8292/© 2021 The Author(s). Published by Elsevier Ltd. This is an open access article under the CC BY license (<http://creativecommons.org/licenses/by/4.0/>).



# Where should parkrun events be located?

**Public Health** 103 (2020) 40–53

Contents lists available at ScienceDirect

**BEST PUBLIC HEALTH**

**Journal homepage:** [www.elsevier.com/locate/phuck](http://www.elsevier.com/locate/phuck)

**Original Research**

**Multiple deprivation and geographic distance to community physical activity events — achieving equitable access to parkrun in England**

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**ARTICLE INFO**

**Objectives:** To evaluate geographic access to free weekly outdoor physical activity events ('parkrun') in England, to identify potential locations for deprived communities, and to identify optimal locations for future parkrun events to further increase access.

**Study design:** This study is a cross-sectional ecological analysis of the socio-economic disparities in parkrun access across England.

**Setting:** All English Lower Layer Super Output Areas (LSOAs).

**Methods:** We combined prediction data on all English Lower Layer Super Output Areas and parkrun events to calculate predicted distances to the nearest event for more than 32,000 communities in England. We used the index of multiple deprivation to measure the degree of socio-economic deprivation, measured using the index of multiple deprivation. We then used geographic information system spatial analysis to conduct a simple location-allocation analysis to identify optimal locations for parkrun events.

**Results:** In England, 93% of the population live within 5 km of one of the 405 parkrun events. There is a significant positive correlation between the number of parkrun events and the index of multiple deprivation. Setting up an additional 200 events in optimal locations would improve average access to the nearest parkrun event by 122 km, from 10.5 km to 9.3 km, and would increase the number of people living within 5 km of a parkrun event by 100,000.

**Conclusion:** Over two-thirds of the English population live within 5 km of a parkrun event, and contrary to common expectation, we find that geographic access is slightly better for those living in more deprived communities. Creating additional events may improve geographic access, but effective strategies will still be required to increase engagement in new and existing events by those living in socio-economically deprived areas.

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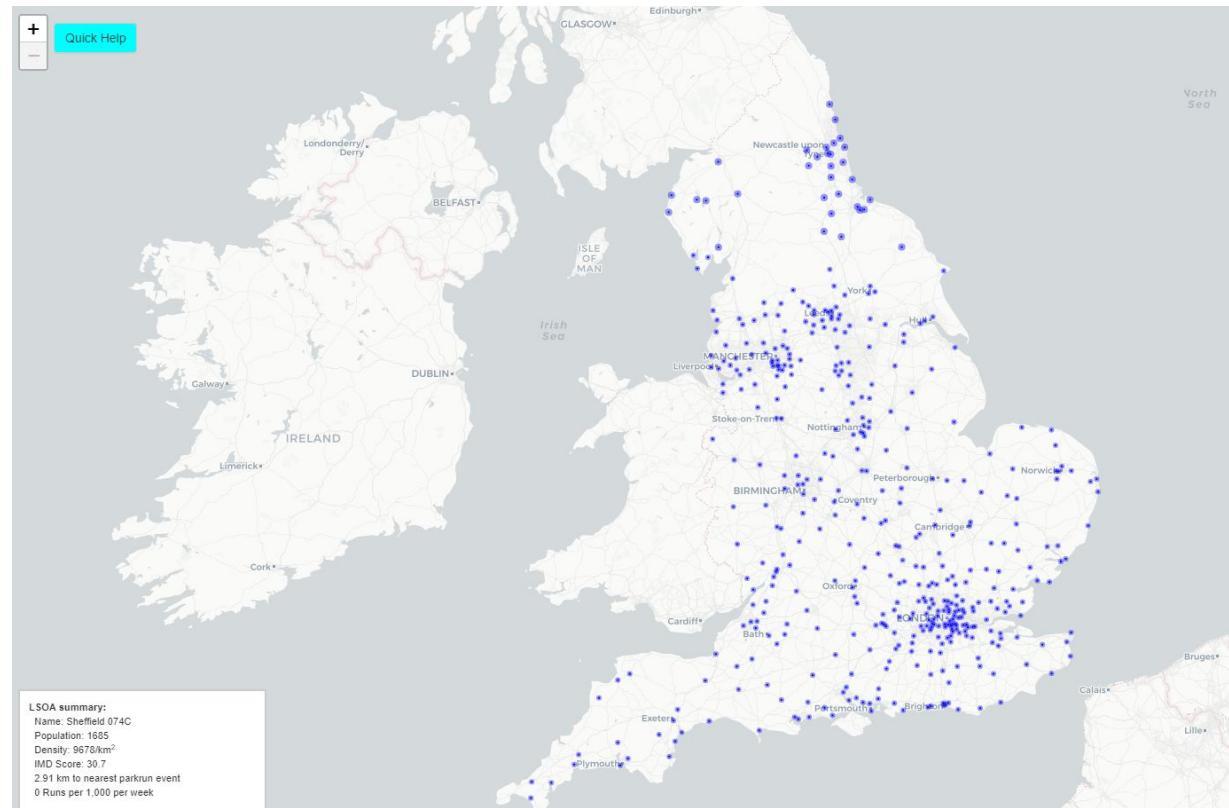
<https://doi.org/10.1016/j.phuck.2020.03.002>  
© 2020 The Author(s). Published by Elsevier Ltd on behalf of The Royal Society for Public Health. This is an open access article under the CC BY license (<http://creativecommons.org/licenses/by/4.0/>).

**Introduction**

Inufficient physical activity is one of the leading causes of disease and disability worldwide.<sup>1</sup> In the UK, around one in six adults do not meet the recommended levels of activity.<sup>2</sup> Activity is also a major contributor to health inequalities, as people from low socio-economic backgrounds are both disproportionately likely to be

inactive<sup>3,4</sup> and be affected by physical inactivity-related diseases,<sup>5</sup> increasing the physical activity levels of the population is therefore high on the public health agenda. It not only has the potential to improve quality of life, reduce mortality rates and alleviate the burden on social services but also reduce the gap in health inequalities.<sup>6</sup>

However, designing effective health interventions to increase physical activity is a considerable challenge.<sup>7,8</sup> Implementing such interventions in a way that does not increase health inequalities might even be more difficult. Studies have shown that programmes to increase physical activity often fail to reach deprived communities and those most in need, suggesting





# Greedy search algorithm

More formally, we define that for any candidate green space location  $j$ , the objective function  $f(j|E)$  provides the sum of parkrun runs  $r_i$  over all LSOA  $i$ , weighted by the squared IMD score  $w_i^2$ , given the set of established parkrun event locations  $E = \{e_1, e_2, \dots, e_{455}\}$ :

$$f(j|E) = \sum_{i=1}^{32844} w_i^2 * r_{ij}$$

In the absence of causal estimates, we use the Poisson regression model specified above to predict the expected number of runs  $r_{ij}$  for LSOA  $i$  based on its IMD score  $w_i$ , its (linear) distance to the nearest parkrun event  $d_{ij}$ , and its population  $p_i$ . The functional form is given below.

$$E(r_{ij}|w_i, d_{ij}, p_i) = \exp(\beta_0 + \beta_1 * w_i + \beta_2 * d_{ij} + \ln(p_i) + \epsilon)$$

Filling-in the parameter coefficients (see table 3), we derive the following formula:

$$\hat{r}_{ij} = \exp(-5.402 - 0.048 * w_i - 0.082 * d_{ij} + \ln(p_i))$$

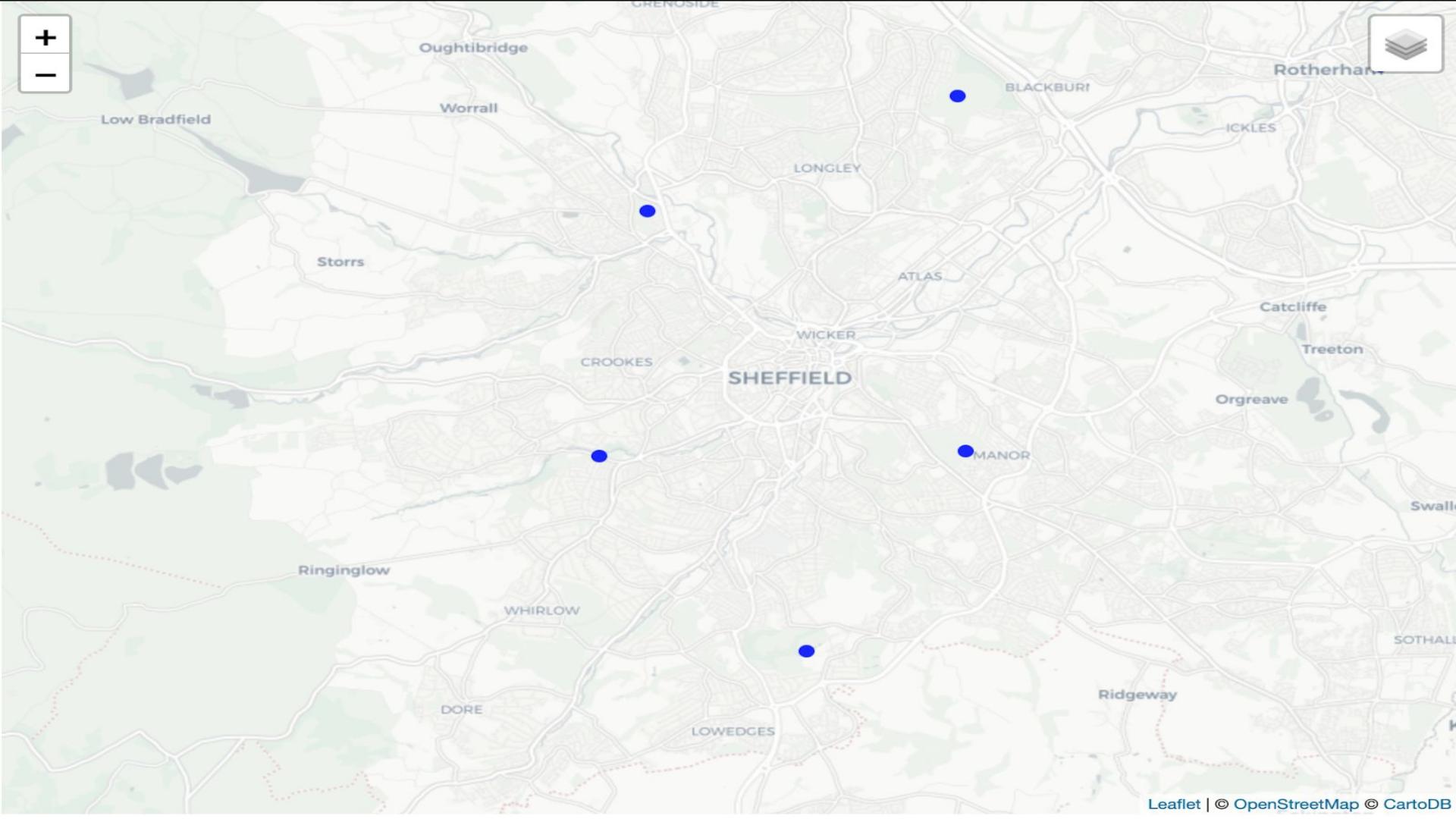
Note that  $j$  can have an effect on  $r_{ij}$  through  $d_{ij}$ : setting up a new event at location  $j$  will reduce the distance to the nearest event for some LSOA  $i$ . This means, we evaluate the distances from LSOA  $i$  to all established parkrun event locations  $\{e_1, e_2, \dots, e_{455}\} \in E$ , denoted  $\overline{l_i e_1}, \overline{l_i e_2}, \dots, \overline{l_i e_{455}}$ , and to the candidate green space location  $j$ , denoted  $\overline{l_i j}$ , and then take the minimum value, i.e.  $d_{ij} = \min(\overline{l_i j}, \overline{l_i e_1}, \overline{l_i e_2}, \dots, \overline{l_i e_{455}})$ .

The expected change in the objective function is computed for all candidate locations  $j$  in the set of the available green spaces  $C = \{c_1, c_2, \dots, c_{2842}\}$ , and the location with the maximum value is selected. The selection function is expressed in the following formula:

$$\arg \max_{j \in C} f(j|E)$$

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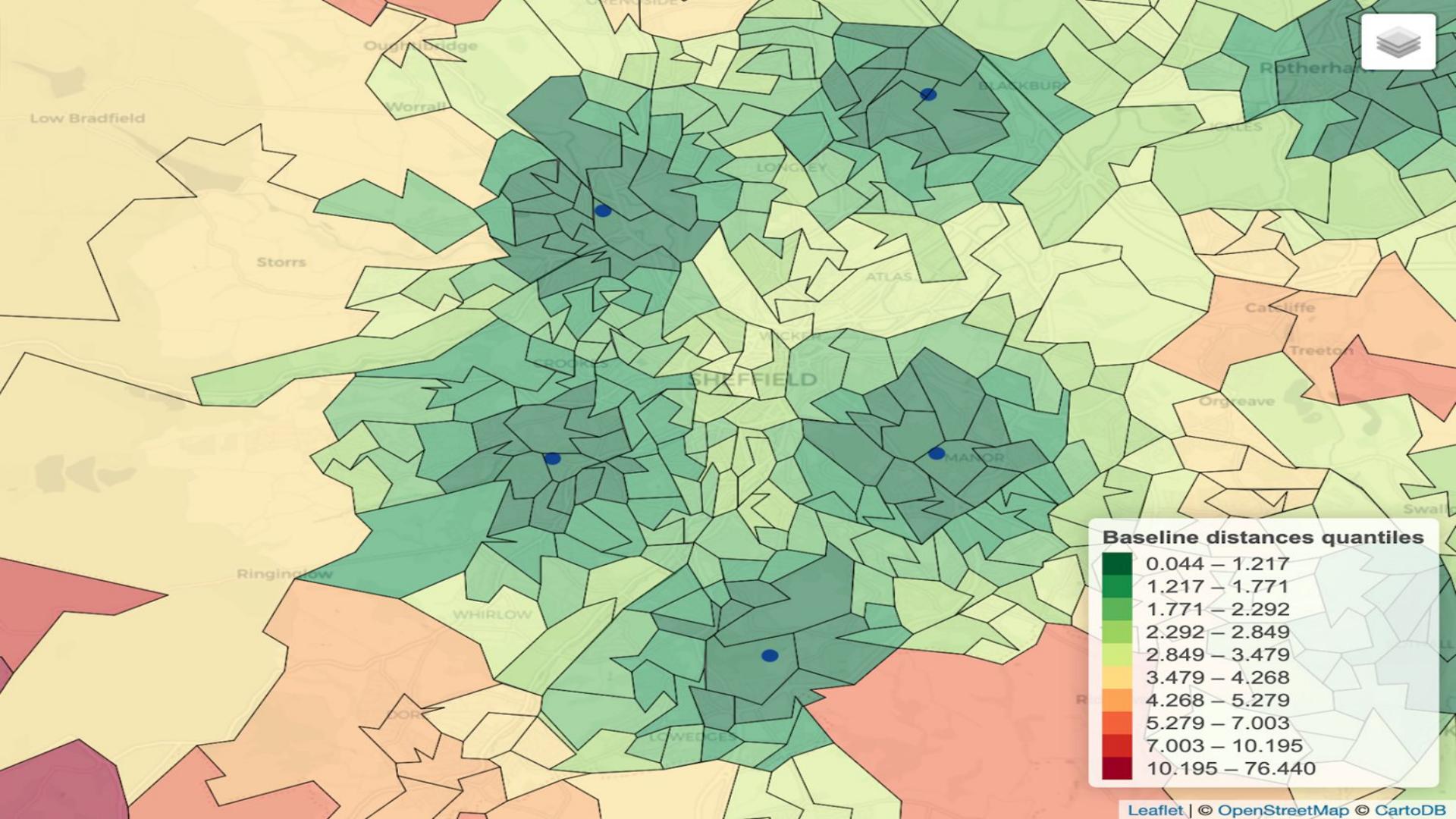
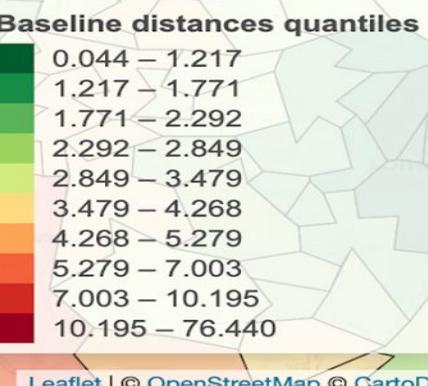


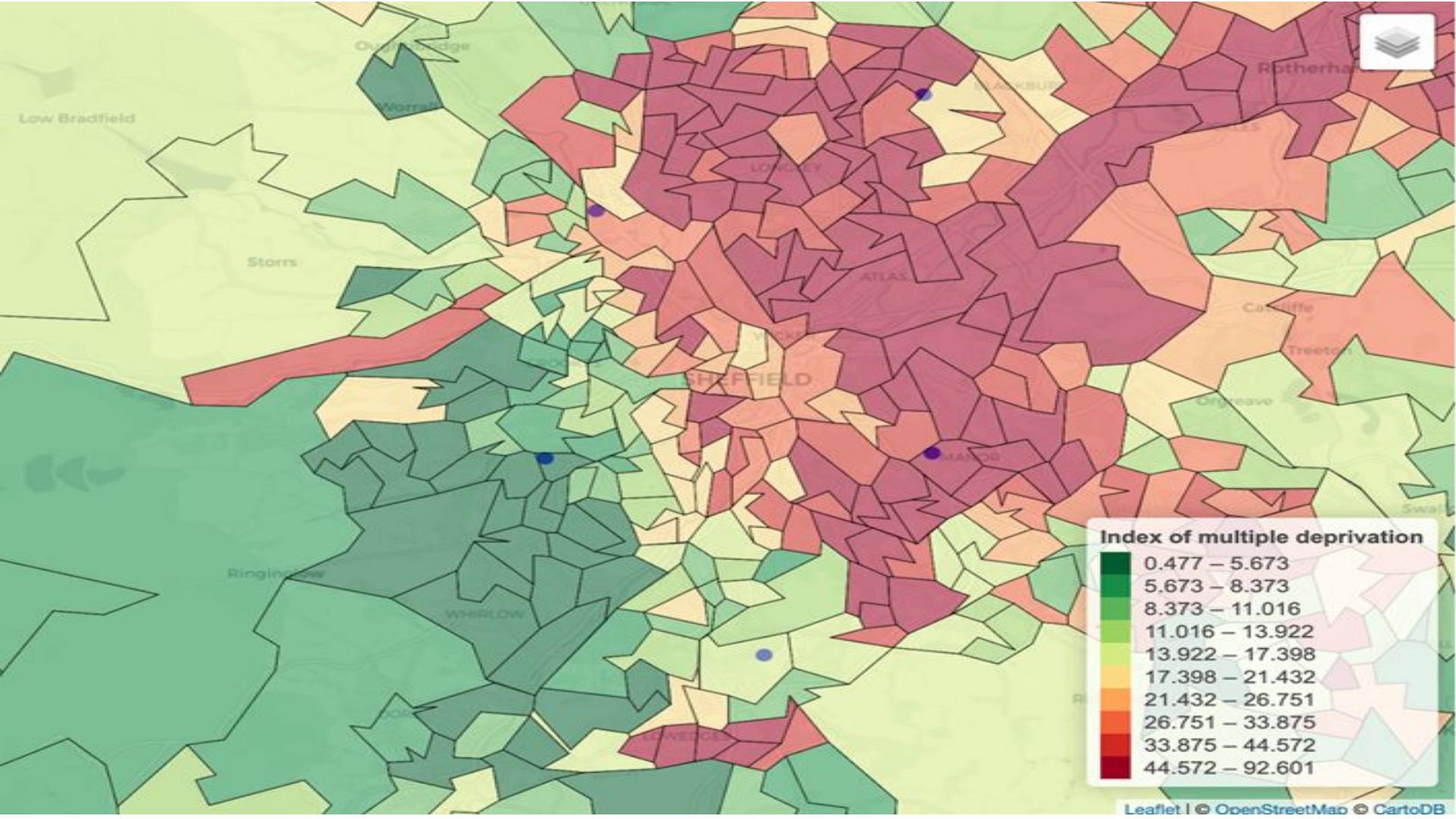


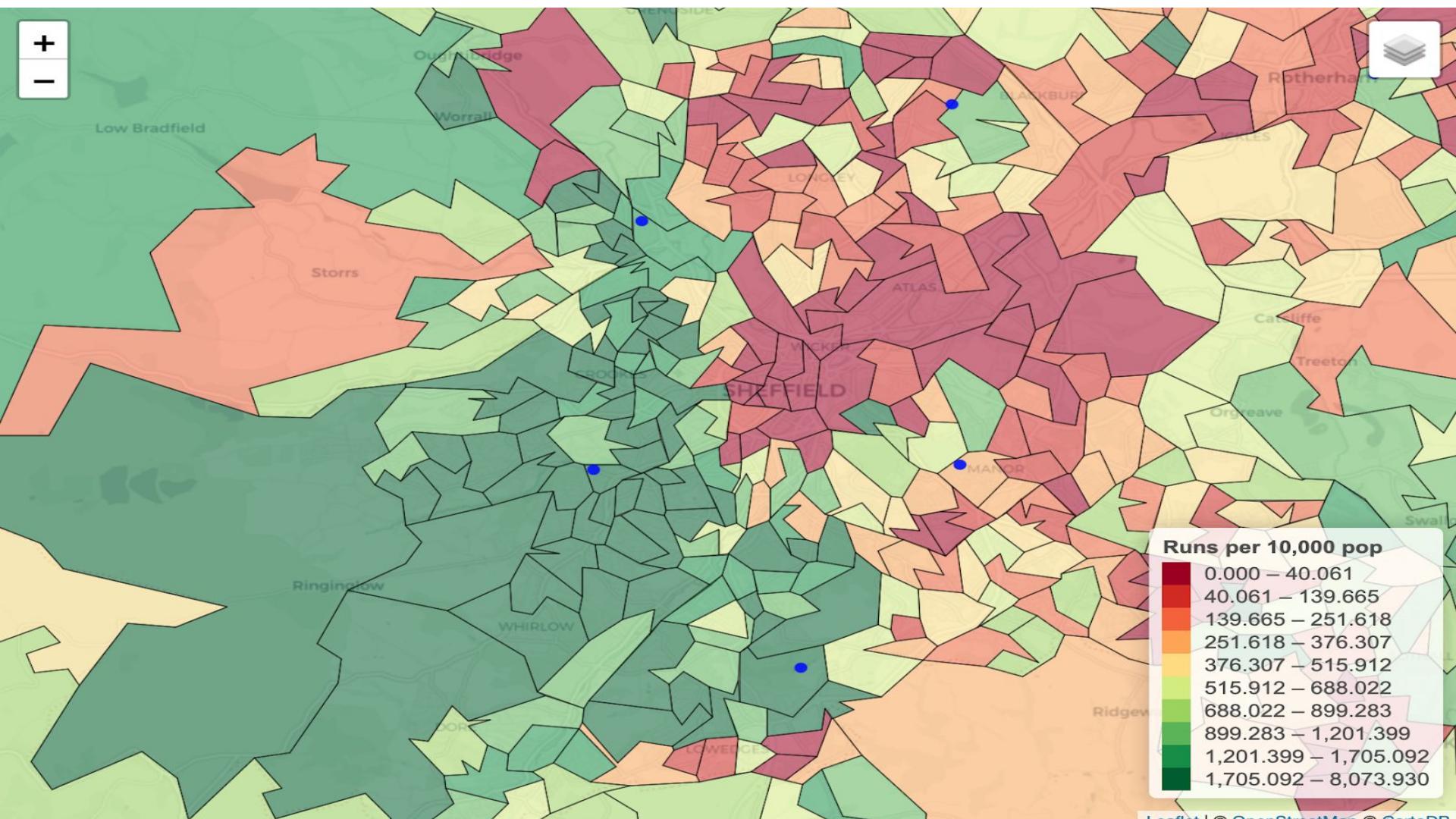
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$$\text{Access} = \{ 2.2 * 1,220 + 4.9 * 2,351 + 2.3 * 1,915 + 0.7 * 1,844 + 2.0 * 1,530 \} = \underline{\underline{22,959.2}}$$







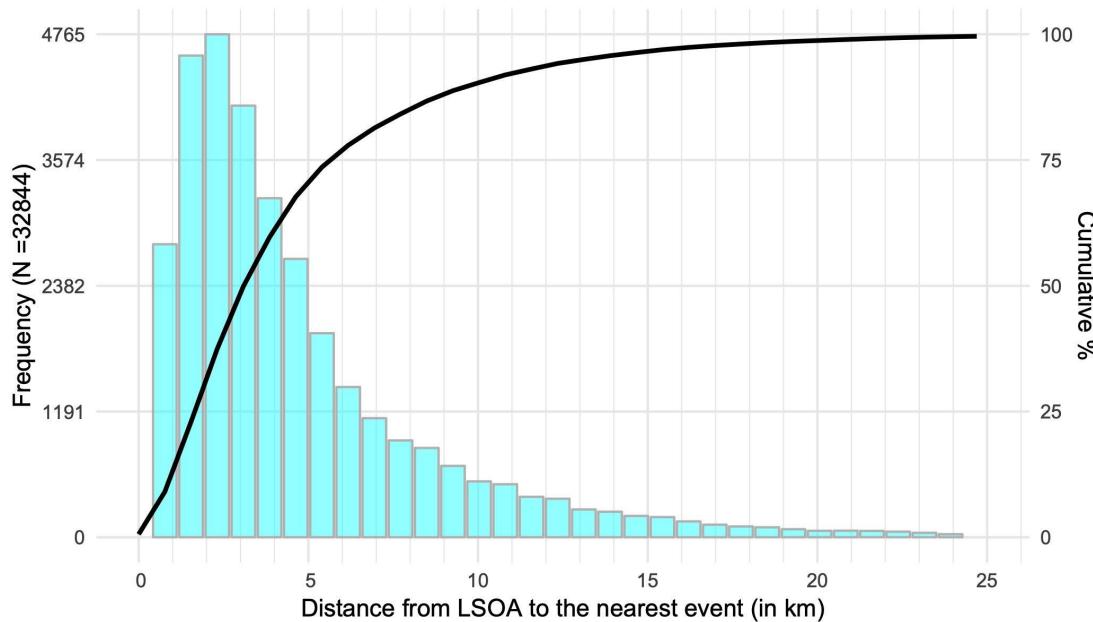
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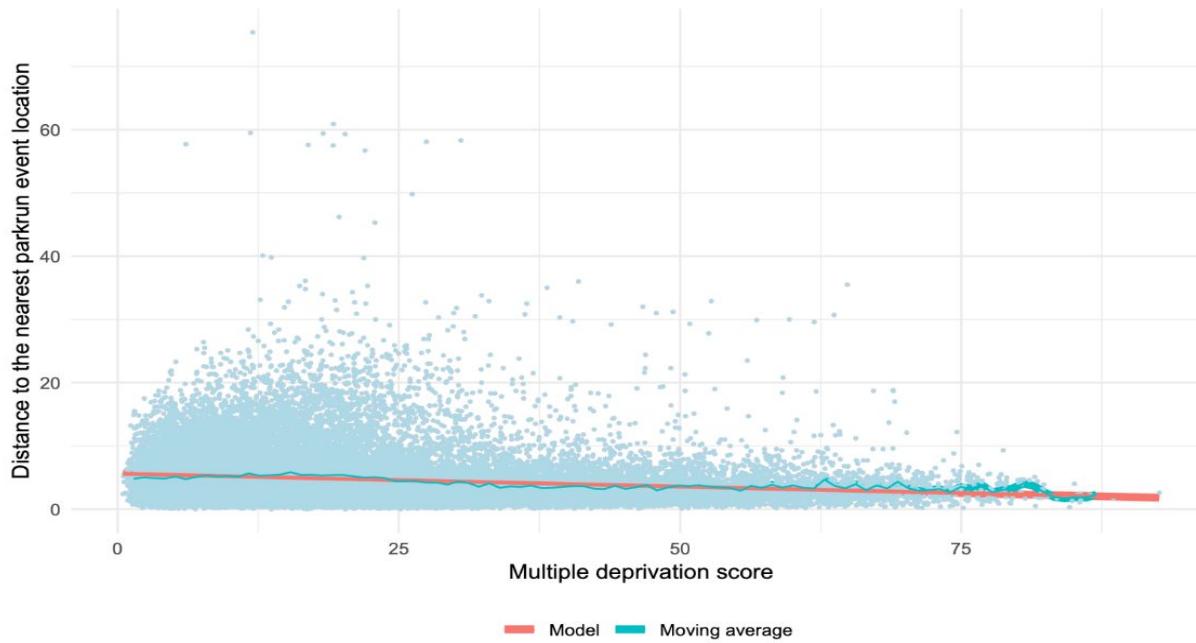
# Results

## 1. How good is geographical access to parkrun?



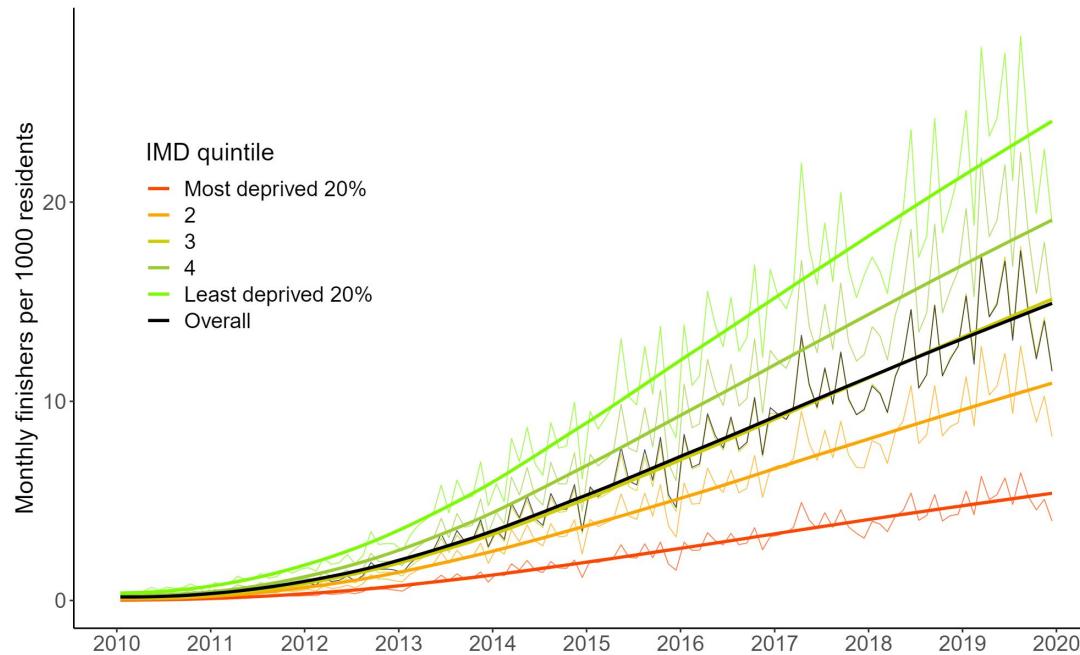
# Results

## 2. How equitable is geographical access?



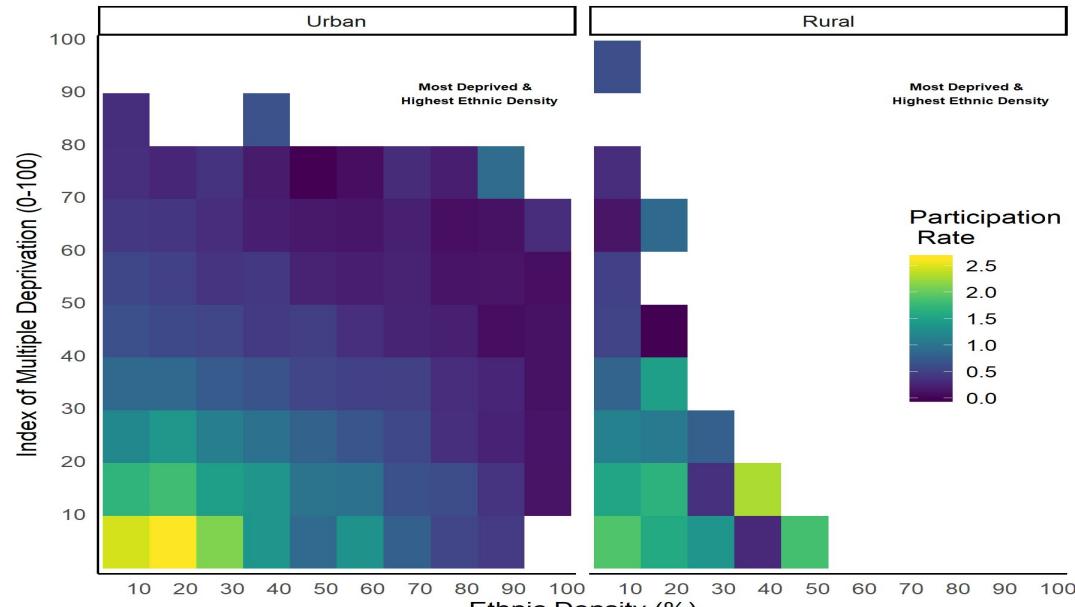
# Results

## 3. How good is participation in parkrun?



# Results

## 4. How equitable is participation?



Sources: Office for National Statistics  
and parkrunUK

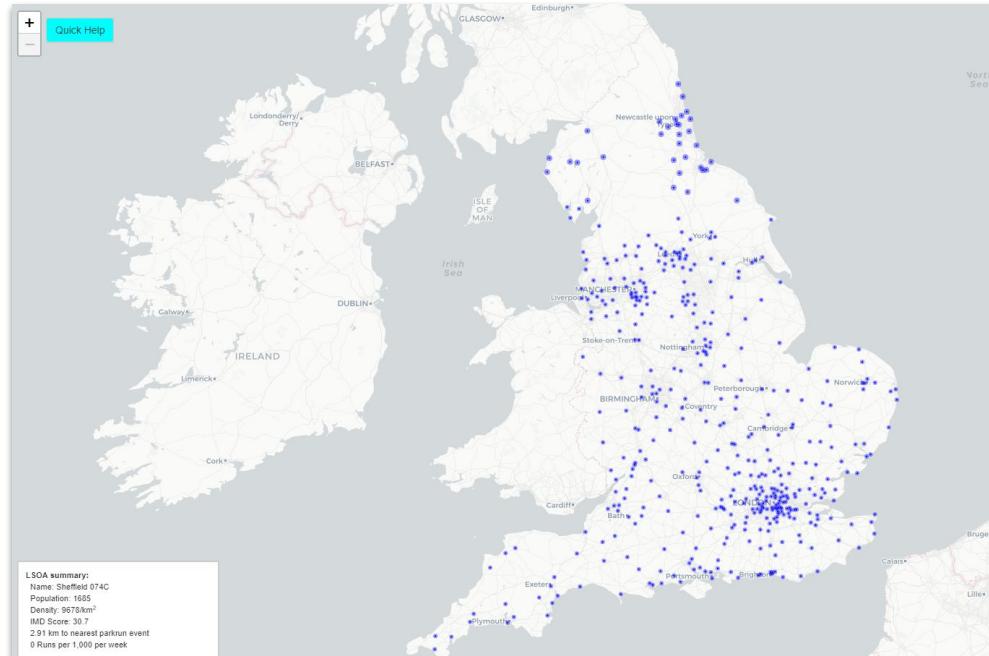
# Where should parkrun locate 200 new parkrun events?



	Access	Participation
Efficiency	1. Maximize overall access.	3. Maximize overall participation.
Equity	2. Maximize deprivation weighted access.	4. Maximize deprivation weighted participation.

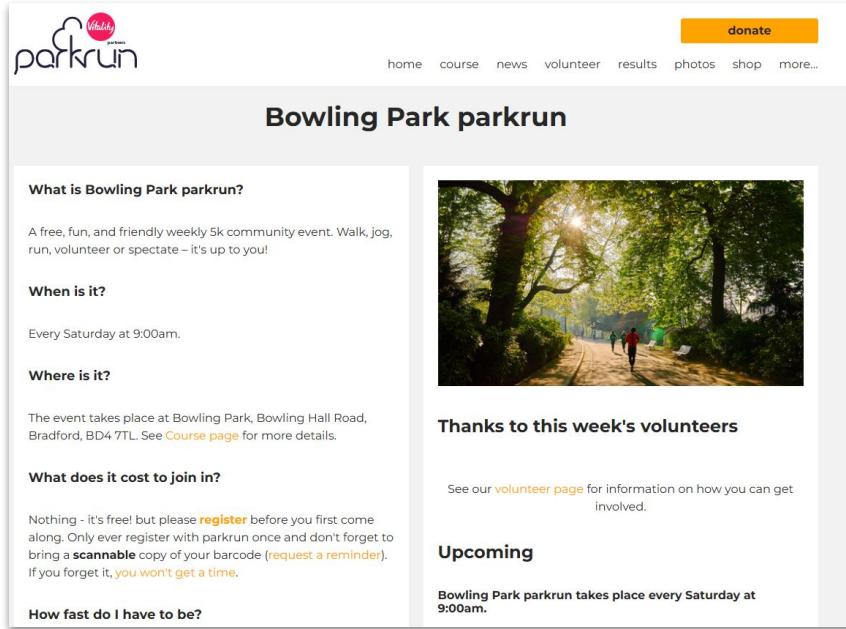
# Results

Where should parkrun locate 200 new parkrun events?



<https://iolmap.shinyapps.io/parkrun/>

# Open source & impact



The screenshot shows the homepage of the Bowling Park parkrun website. At the top, there's a navigation bar with links for home, course, news, volunteer, results, photos, shop, and more... A yellow 'donate' button is also visible. Below the navigation, the title 'Bowling Park parkrun' is displayed. To the left, there's a section titled 'What is Bowling Park parkrun?' followed by a brief description of the event. To the right, there's a large photograph of a park path with people jogging. Below the photo, there's a section titled 'Thanks to this week's volunteers' with a link to the volunteer page. Further down, there's an 'Upcoming' section stating that the event takes place every Saturday at 9:00am.

**What is Bowling Park parkrun?**

A free, fun, and friendly weekly 5k community event. Walk, jog, run, volunteer or spectate – it's up to you!

**When is it?**

Every Saturday at 9:00am.

**Where is it?**

The event takes place at Bowling Park, Bowling Hall Road, Bradford, BD4 7TL. See [Course page](#) for more details.

**What does it cost to join in?**

Nothing - it's free! but please [register](#) before you first come along. Only ever register with parkrun once and don't forget to bring a [scannable](#) copy of your barcode ([request a reminder](#)). If you forget it, [you won't get a time](#).

**How fast do I have to be?**

*“Rob and colleague Dr. Paul Schneider developed a statistical tool (an algorithm) which searched through all of the greenspaces in England and ranked the top 200 by predicted public health impact.*

*One example of how the statistical tool was used is the creation of Bowling Park parkrun, located in a deprived area of Bradford. Our local Ambassador, working with community groups, identified the location as an option for a parkrun event – which was corroborated by Rob’s work – and the event became a reality for the local people.” (parkrunUK, 2020).*

<https://blog.parkrun.com/uk/2020/12/08/using-research-to-improve-inclusivity/>

# Open source

## Does ethnic density influence community participation in mass participation physical activity events?

Authors: Robert A. Smith, Paul P. Schneider, Alice Bullas, Steve Haake, Helen Quirk, Rami Cosulich, Elizabeth Goyder

DOI: 10.12688/wellcomeopenres.15657.2

Submitted by rasmith3

Mean reproducibility score: 9.2/10 |  Number of reviews: 5

<https://www.reprohack.org/paper/28/>

## Where should new parkrun events be located? Modelling the potential impact of 200 new events on socio-economic inequalities in access and participation

Authors: Schneider PP, Smith RA, Bullas AM, Bayley T, Haake SS, Brennan A, Goyder E

Submitted by hub-admin

Mean reproducibility score: 7.0/10 |  Number of reviews: 3

<https://www.reprohack.org/paper/79/>



Fran Biggin  
@francesbiggin

247  
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Having a great time at the @N8CIR ReproHack reproducing the research in this really interesting PrePrint about @parkrunUK by @waq0r, @R06ertSm1th and colleagues:

Smith et al. BMC Public Health (2022) 22:1542  
<https://doi.org/10.1186/s12889-022-13981-5>

BMC Public Health

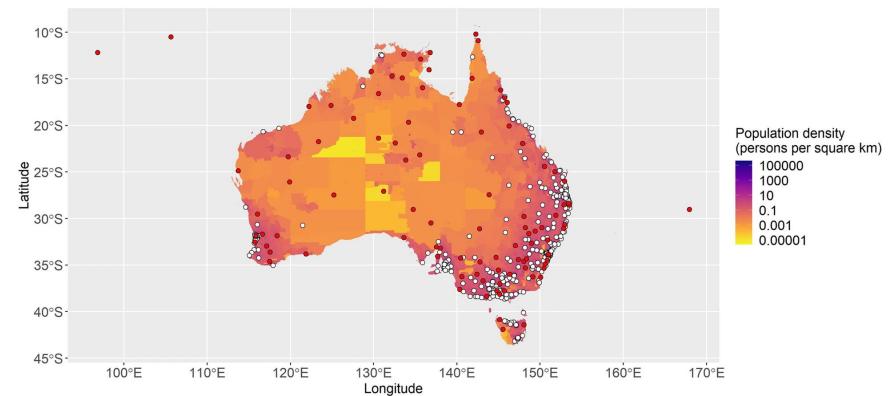
RESEARCH

Open Access



## Improving access to public physical activity events for disadvantaged communities in Australia

Janette L. Smith<sup>1,2</sup>, Lindsey J. Reece<sup>1</sup>, Catriona L. Rose<sup>1</sup> and Katherine B. Owen<sup>1\*</sup>





# Impact of the pandemic on parkrun

# Pandemic Policy on 'parkrun' Participation

Rousham et al. BMC Public Health (2024) 24:2931  
 https://doi.org/10.1186/s12889-024-20420-0

**BMC Public Health**

**RESEARCH** **Open Access** 

## The long-term effect of the coronavirus pandemic on parkrun participation: an interrupted time series analysis

Oscar Rousham<sup>1</sup> , Helen Quirk<sup>1</sup> , Elizabeth Goyer<sup>1</sup>  and Robert A. Smith<sup>1</sup> 

**Abstract**

**Background** The growth of parkrun between 2004 and 2019 has been heralded as a success story for public health as a result of its physical activity and wellbeing benefits for participants. However, parkrun was not immune from the COVID-19 pandemic – with events in mainland England cancelled from March 2020 to July 2021. This study explores the lasting impact of the pandemic on parkrun participation to February 2023, and its implications across the socioeconomic spectrum.

**Methods** The study combines aggregated parkrun weekly finisher data from 32,470 Lower Layer Super Output Areas (LSOA) in England from January 2015 to February 2023 with Office of National Statistics (ONS) data on population and deprivation. Interrupted time series analysis using segmented Poisson regression models was used to estimate the immediate change in parkrun participation and the change in the rate of growth following the pandemic. Models were fitted for each Index of Multiple Deprivation (IMD) quintile separately to assess whether this effect differed by socioeconomic deprivation.

**Results** Visualisation and interrupted time series analysis showed a significant and long-term decrease in parkrun participation following the reopening of parkrun events. This was consistent across all IMD quintiles, indicating that the inequalities in parkrun participation according to IMD observed prior to the pandemic remained after the pandemic. Between March 2020 and February 2023, almost 13 million fewer parkrun finishes are estimated to have occurred relative to what would have occurred in the absence of the pandemic.

**Conclusion** The reduction in parkrun participation during the pandemic and following the reopening of events is likely to have negatively impacted wellbeing in would-be participants. Going forwards, policymakers must make the difficult trade-off between the long-term health and social implications of restricting outdoor physical activity events against the benefits associated with a reduction in infectious disease transmission.

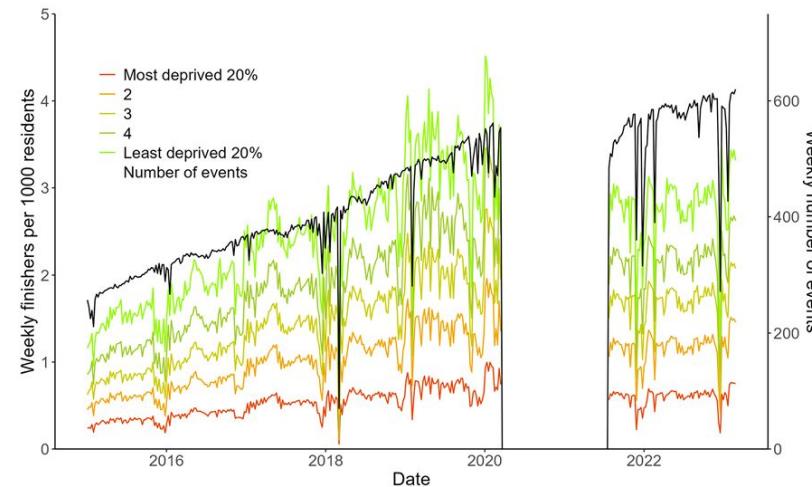
**Keywords** Parkrun, Physical activity, Socioeconomic deprivation, Ecological study, Interrupted time series

**Introduction**

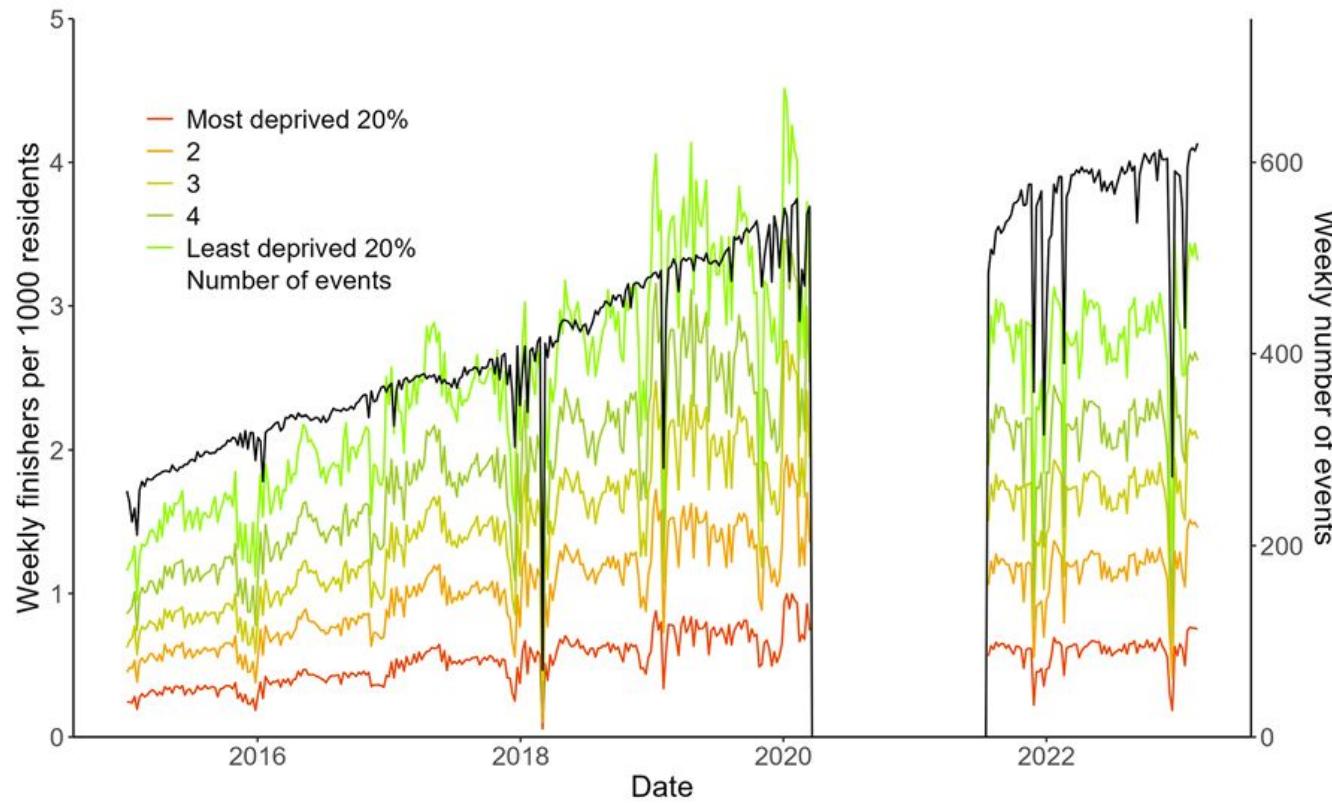
Engaging in regular physical activity is linked to a decreased risk of developing numerous non-communicable diseases [1], along with notable reductions in depression and anxiety [2]. However, a significant portion of the population falls short of recommended activity levels [3] and there is socioeconomic inequality in leisure time physical activity levels [1, 3, 19]. Elevating

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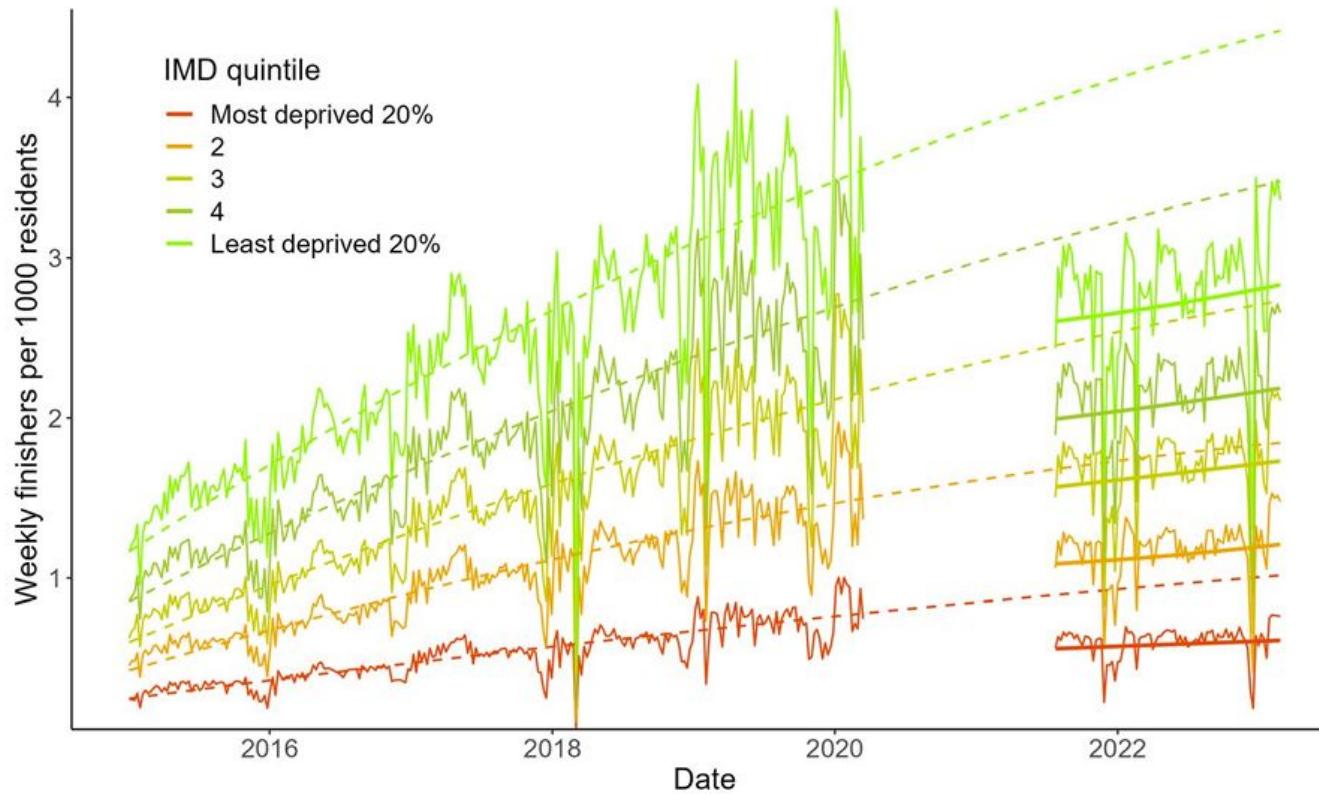
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**Fig. 1** Weekly number of parkrun finishers in England per 1,000 residents by Index of Multiple Deprivation Quintile, and number of parkrun events in operation, from January 2015 to February 2023



**Fig. 1** Weekly number of parkrun finishers in England per 1,000 residents by Index of Multiple Deprivation Quintile, and number of parkrun events in operation, from January 2015 to February 2023



**Fig. 2** Counterfactual (expected participation in the absence of the pandemic) compared to observed participation. The thin lines show the observed participation rates by IMD each week. The dotted lines show the pre-pandemic trend ignoring seasonal variation extrapolated across the study period. The solid lines show the post-pandemic trend, ignoring seasonal variation



# Publications

Smith, R.A., Schneider, P.P., Cosulich, R., Quirk, H., Bullas, A.M., Haake, S.J. and Goyder, E., 2021. Socioeconomic inequalities in distance to and participation in a community-based running and walking activity: A longitudinal ecological study of parkrun 2010 to 2019. *Health & Place*, 71, p.102626.  
<https://doi.org/10.1016/j.healthplace.2021.102626>

Schneider, P.P., Smith, R.A., Bullas, A.M., Bayley, T., Haake, S.S., Brennan, A. and Goyder, E. 2020. Multiple deprivation and geographic distance to community physical activity events — achieving equitable access to parkrun in England. *Journal of Public Health*. 48;53(189). <https://doi.org/10.1016/j.puhe.2020.09.002>

Smith, R., Schneider, P., Bullas, A., Haake, S., Quirk, H., Cosulich, R. and Goyder, E., 2020. Does ethnic density influence community participation in mass participation physical activity events? The case of parkrun in England. *Wellcome Open Research*, 5(9), p.9. <https://doi.org/10.12688/wellcomeopenres.15657.2>

Rousham, O., Quirk, H., Goyder, E. et al. The long-term effect of the coronavirus pandemic on parkrun participation: an interrupted time series analysis. *BMC Public Health* 24, 2931 (2024).  
<https://doi.org/10.1186/s12889-024-20420-0>



# Open Source Code

[https://github.com/RobertASmith/DoPE\\_Public](https://github.com/RobertASmith/DoPE_Public)

[https://github.com/RobertASmith/parkrun\\_temporal\\_23](https://github.com/RobertASmith/parkrun_temporal_23)

[https://github.com/RobertASmith/parkrun\\_temporal](https://github.com/RobertASmith/parkrun_temporal)

<https://github.com/RobertASmith/parkruntimeseries>

[https://github.com/ScHARR-PHEDS/iolmap\\_revision](https://github.com/ScHARR-PHEDS/iolmap_revision)

[https://github.com/ScHARR-PHEDS/parkrun\\_book](https://github.com/ScHARR-PHEDS/parkrun_book)



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## 1 Using this book

**Note from the authors:** This work is a living document and is being adapted all the time based upon comments. New sections are being included prior to teaching. If you have suggestions for improving this book, please contact Robert Smith by email: [rsmith@darkpeakanalytics.com](mailto:rsmith@darkpeakanalytics.com).

— Dark Peak Analytics Teaching Team

This book was created to provide additional support for the taught course. It includes all of the code chunks, exercises and solutions which we cover in the taught sessions. It serves as a first point of reference, and directs the reader to additional resources. The book is written in R using the `bookdown` package, which converts each chapter from an RMarkdown file into a PDF or HTML book. Later in the course, we will cover how this process works.

There are currently thirteen chapters in the book, as shown in the table below. The content that is visible to you will depend on the scope of the course that you are enrolled on. For example those taking a course on the use of shiny for Cohort State Transition Models may have access to Chapter 7 and 10, whereas those doing an introductory R course may have access to chapter 1 only.